

Use of Communication Ranging for Optimizing the Localization Accuracy of Mobile Sensor Networks

Raxit G. Jani^a, Vishesh Joshi^{b*}, Priyank Bhanushali^c, Manav Nai^d

^aAssistant Professor, Computer Engineering Department, Gandhinagar Institute of Technology, Gandhinagar 382721, Gujarat, India

^{b*,c,d}Student, Gandhinagar Institute of Technology, Gandhinagar 382721, Gujarat, India

Abstract

Localization is among the most important tasks performed by sensor networks. Wireless sensor network (WSN) functions like coverage and event detection are affected by the localization method that is utilized. Accuracy is considered to be among the factors that has been connected to good network efficiency in the most of applications utilizing sensor networks. Communication ranging is one easily implementable technique for positioning on short transmission nodes without any additional elements. The work suggests a highly accurate improved communication ranging localization technique. Twenty Wasp mote Nodes and a Meshlium router were used to test and validate the suggested technique in a forest environment. Uniqueness of suggested localization method is centered upon adjusting the XBee transmitter's capacity to change power over four rounds in order to deliver the best location prediction. The findings showed that nodes' locations may be determined with an error range of 0 to 22 metres. Starting at 30 in round 1, the network location inaccuracy dropped to 8 metres in round 4. This method is utilized with a variety of networks and technologies as far as the system capacity could be adjusted to distinct values and the transmission range is determined or can be manually evaluated.

1. Introduction

Wireless sensing technologies have seen exceptional progress in recent years, and a different type of wireless sensor network system has arisen[1]. Moreover, people's desire for associated technologies is rising daily. This technology is widely used in the army, commerce, agriculture, raising livestock, home automation, smart grid, and numerous other sectors because of its minimal price, small footprint, with straightforward distribution of IOT devices[2]. Although each of these applications is diverse and has its own unique features, they all require the sensor nodes' locations in order to function [3]. All of the system's information becomes meaningless if the sensor node's location information is absent. The method for obtaining node position information is so crucial. For a very long period, GPS positioning devices have been used to acquire sensor location data [4]. We know that using Global positioning technology is expensive. Our specifications for low-cost sensor nodes are not met by this. How to control the expense of finding sensor nodes is currently a prominent topic [5].

Whenever wireless sensor networks are implemented in enclosed spaces and urban areas, the data transfer among sensors may frequently get hindered via objects, preventing the device from communicating in alignment. Such a problem has two direct effects: it significantly alters the data they obtain from the locating module and reduces the locating performance of the system[6]. In order to resolve these problems, this paper proposes a nonline-of-sight identification technique. A search algorithm finds the sensor which can interact with the system via connection and eliminates the node throughout positional play, significantly altering some of the information with the nonline-of-sight feature and removing the node during positioning, removing the impact of the unaligned line of sight issue on structure placement efficiency.

Throughout this paper, biochemical gas sources are located using wireless sensor networks to aid individuals in responding to disasters more quickly. We propose a robust expectation - maximization location technique and compare it with the direct trilateral approach as well as the nonlinear least squares approach. In order to address imperfections of both the centralised location algorithm and preserve network bandwidth, the multilayered sensor network has been implemented on biochemical resource placing. A distributed positioning algorithm based on improved particle filtering within the hierarchical sensor network is then proposed, which manages nodes through sub regions. State changes, and the convergent biochemical gas source position is estimated iteratively in a loop. The creation of a brand-new sensor node location information gathering technology is a top priority for many R&D organisations. Currently, it is common practice to rely only on a wireless sensor network for locating. This technology has the unique property that, in order to complete positioning, It merely requires using information from interaction between nodes or an exclusive sequence of instructions.. In order to examine the present issues with the technology and provide solutions, this article introduces several positioning algorithms, sensor network positioning technologies, and associated topics.

2. Related Work

The localization can be divided into localizations based on known locations, proximity, angles, ranges, and distances. Target localization and node self-localization were the two categories into which Cheng et al. (2012) divided the known localization

*Vishesh Joshi

E-mail address: visheshjoshi78@gmail.com

techniques [10] for their review. Their localization method taxonomy is displayed in the following figure (see Fig. 1).

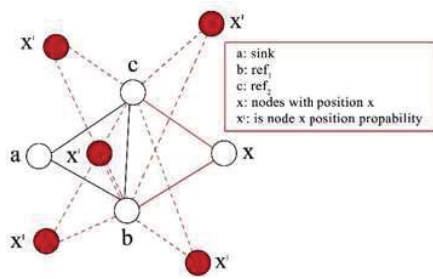


Fig. 1. Angle calculation.

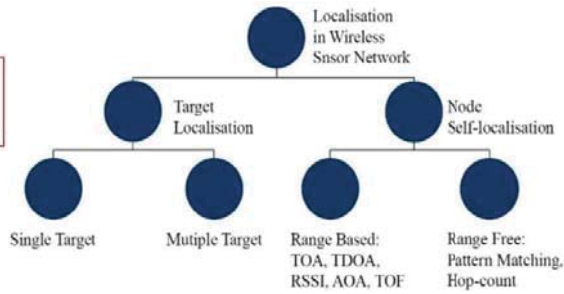


Fig. 2. Localization method taxonomy.

The far more dependable and promising navigation innovations include thought to be GPS or standalone cellular systems [11][12]. Although GPS is significant because to its widespread accessibility, it is impracticable to place it inside every sensor due to its high cost and energy consumption. As a consequence, in order to conserve power and resources, just a tiny subset of nodes, referred to as the anchor as well as beacon nodes, carry GPS units. With one of the several localization techniques, the remaining nodes learn their positions. Determination of reception phase of the antennas, the AoA technique compares the beam shapes and signal intensity. To perform this comparison, this tactic needs both directional and omni-directional antennas. A precise synchronisation is necessary for either the round or one-way propagation period to be provided by distance-related measurements. The signals from various receivers are measured, and the measurement is then used in some calculations, like cross-correlation, to determine the synchronized approximation of the transmitter provided by the TDoA. Without taking angle into account, the RSSI determines its output intensity in relation to the transmitter and receiver's range. Essentially, there are an endless amount of disturbances intermingled with the channel power, but RSSI measures energy on the channel over time.

A biased location estimation vector had to be created for Yu to introduce an AoA-ToA transmission hypothesis in 2007[11]. Khan et al. [26] far nearly presented an enhanced and objective rendition in which they carried out a simulation to test whether the performance of the linear least squares estimator might be reliably predicted using the theoretical mean square error. An approximation pertaining towards the challenge of quasi target localization in a 3-D space serves as the foundation for the LLS and its optimisation, a different method Yu proposed in 2007 [27]. Khan et al. [28] presented a similar strategy two years later, with the primary exception being that the masses included the path-loss coefficient was utilized in their approach, and treated as an undetermined figure. Jiang et al. [14] established a slant-based localization model where a number of RSSI lights were used to establish the inclinations where technique utilizing biometrics as a foundation cease to deliver adequately when the network's circumstance altered. While Tomic et al. additionally established AoA.[29], the localization approach employed the RSSI. Lacking a description of the dissemination state where the Predicting AoA was regarded crucial, the nodes in these span approaches were located by the AoA. If distances are known, A methodology to evaluate the position inclinations was put out by Alkhatib et al. [30]. This method is based on the polar system's (R &) ability to localize nodes (see Fig. 2). The node spacing are contrasted (R) from unidentified and recognised nodes, the angles were estimated.

The localization approach was divided into centralised and dispersed techniques by Paul and Sato [31]. According to them, For WSN apps, the dispersed strategy was more commonly employed. We also noted that despite free ranging methodologies were a greater expense in terms of node hardware, ranging methodology were thought to be quite precise. A mathematical formulation that is variable for WSN localization by Liu and Liu [32] was put forth using measurements of the separations between M nodes in M-dimensional space. To determine the radial error, nodes with comparative weights were stored in a matrix.

3. Analysis of Available Techniques

Precise localization of sensors is among the very important criteria inside this plurality of data networks. Unfortunately, every possible approach to achieve this requires extra technology, like GPS units, sonar or thermal sensors, beams, or indeed directed radar. Each of these options raises the node's material requirements, requiring more power and spending more money on faster memory and CPUs. However, additional tests and experiments revealed that RSSI was still not dependable within circumstances where connections encountered problems like high absorbance or uncertain coefficients of path-loss. Numerous investigations have discovered that RSSI is indeed a node positioning indicator. According to a study by Cama-Pinto [33], RSSI can only be used as an indicator for distances under 40 metres; for distances between 40 and 100 metres, readings can vary between 60 and 65 dBm, which makes it difficult to use as an indicator. In actuality, these values weren't constants because almost everything had an impact on the outcomes, including the environment outside and inside, the weather (dry or humid), shifting elements and obstructions. As a result, RSSI analysis was deemed unreliable or incorrect, or they may produce false findings in few specific softwares. A ToF and ToA are very popular approaches for localization in WSNs and various types of telecommunication systems. The following factors, however, limit the technique's applicability to theoretical studies of short-range communication:

$$\text{Speed} = \frac{\text{Distance}}{\text{Time}}$$

to calculate the required time for 1 meter

$$T = 1 \frac{\text{meter}}{3} * 10^8 = 3.33 \text{ nsec}$$

One meter's difference was measured in 3.33 nsec, and ten meters' difference in 33.3 nsec. In order to reliably predict the ToF or ToA by 1 meter and 30 MHz over a span of ten meters accuracy, It requires a CPU with a speed of at least 300 MHz (1/3.33 nsec). According to sensor network applications, nodes are spread out between each other at a range of 1 to 100 metres in the majority of cases where the sensors' detecting range only extends to a few metres or up to 700 nodes when utilizing an Xbee in the form specified in the Guide for Waspote [34]. Arduino, Waspote, or sensor nodes found on store shelves in general are not able to deliver such accuracy. In Figure 3,

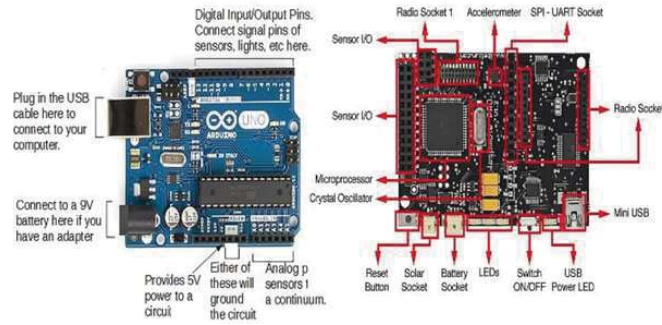


Fig. 3. Arduino and Waspote.

Waspote and Arduino both make use of ATMEGA processors. The ATMEGA 1281 featuring a 14.74MHz bandwidth is utilized by Waspote, which translates to a processor's ability can generate 14,740,000 pulses each second to a clock. Datasheet for something like the ATMEGA1281 [35] indicates that the CPU comprises numerous timers that are utilized for various functions (see Fig. 4). The same source input, known as "AVR Clock," the first instance of this was delivered by a frequency-controlled crystal oscillator of 14.74 MHz, is used by all timers. These timers are used to time when the transmitter and receiver turn on and off. besides other timed activities. These timers have the ability to measure TOF, TOA, etc.

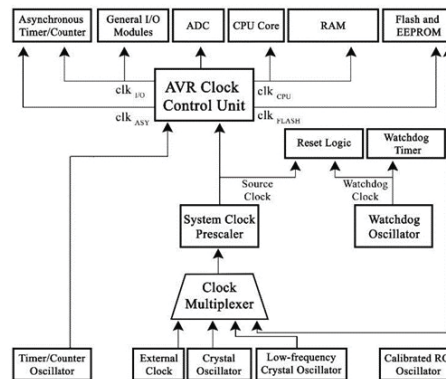


Fig. 4. Clock Distribution. [31]

The duration of a clock pulse is $\text{instr.}484260515 \text{ nsec}$ ($1/f$). Waspote's minimum measurement duration is consequently approximately 67.5 nsec. Thus, $a = 20.27 \text{ m}$ when a Means (67.5 nsec denotes the duration of a single pulse) / (3.33 nsec TOF for 1 meter). Therefore, a clock = 20.27 meters. This technique was tested using Waspote, but it was discovered that the component became unreliable for measuring pulses of the clock. A total of six devices were present in ATMEGA, but not one of them were capable of accurately counting a clock's phases [32] because the actual clock might drift with each trial and because they couldn't distinguish between processing-related pulses and unimportant pulses while counting pulses at such a fast speed. Without utilizing one of the prescaling choices of 64 and above, these timers provide unreliable counting (see Table 1). To put it another way, every count increases by sixty-four instr (or, more specifically, each counter value is increased by sixty-four instr) [36]. As a result, the TOF needed for 1297.3 meters is $64 * 67.5 \text{ nsec}$, or 4320 nsec. Xbee, Zigbee, and other short-term protocols for communication for detector equipment are in fact exempt from this. However, it is limited to being used to Satellite or satellite data, which require millisecond-level timing precision and whose signals travel hundreds of thousands of miles. Waspote can make precise calculations for these signals. The ATMEGA328p microcontroller used by Arduino operates at a frequency of around 20 MHz,

making one clock cycle equal to 50 nanoseconds or a 15-meter TOF. This problem affects Wasmote and Arduino as well, making it challenging to tally pulses of clock at this rate.

4. Methodical Study

Communication-based localization is regarded as a widespread technique. Further research revealed that it was the sole method that could be used on sensor nodes lacking any other parts, like GPS or ultrasonic sensors. We demonstrated an innovative approach using different Wasmote electrical transmission settings to identify the transmission frequencies relative to source networks. Several rounds of this procedure were used to attain improved accuracy. Twenty Wasmote nodes and a way of collecting networks were used in an experimental measurement to determine the proposed method's accuracy. (a Spanish product from Libelium) [33]. (see Fig. 5).



Fig 5. Wasmote and Meshlium. [37]

Wasmote nodes with an XBeePRO-S2 emitter that supports 802.15.4. According to Libelium [37], its transmission range can reach 700 metres. The first among five data can be chosen for the XBee power transmission. We manually measured the communication distance for the values (0, 1, 2, and 3) and presented the results (see Table 2). The experiment used a total of nine nodes as references, with data on their positions (X, Y) kept in memory. The experiment was conducted in a woodland environment (see Fig. 6). (reference nodes). All other 11 nodes were deployed at random and used specially designed frames in conjunction with the Wasmote IDE for locating one another through contacting each other nearby.



Fig. 6. The localization measurement in the forest.

During initial configuration, a variety of referencing sites must be manually loaded. These remaining nodes are randomly distributed (see Fig. 7).

The following is a summary of the suggested technique rounds:

Phase 1

- Because the power levels for every node were all configured with 0 = 10 dBm ways of collecting [xbee802.setPowerLevel(0)], the initial range for transmission and reception was 100 m.
- Reference nodes begin broadcasting messages with the node ID and their coordinates as (X, Y). 5 Every single unfamiliar network attempts to get as many notifications as it's able to receive from in-range nodes that serve as references in order to determine every possible position within the cross-section of every one of the in-range standard networks.

- Two reference nodes must be within range at the very least (see Fig. 9). Alternatively, hold off until more accessible points of reference are available shortly.
- Location of Phase 1 is conceivable $[] = [x1 \ y1 \dots \ xn1 \ yn1]$
- The Phase1 results include all potential sites and the average of all potential nodes as recommended locations for this phase.

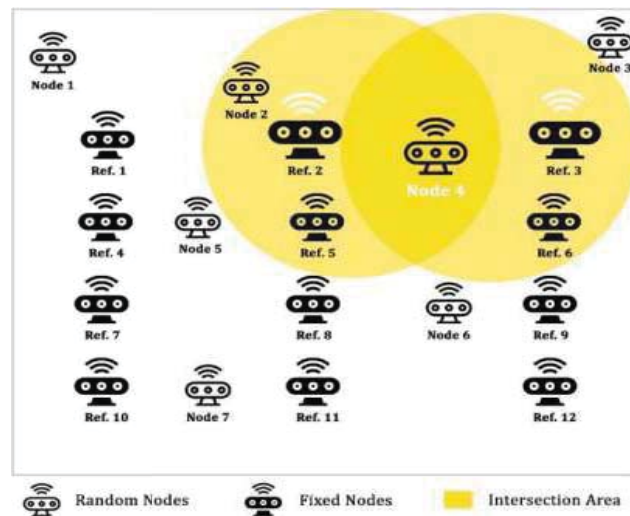


Fig. 7. Potential sites inside the chosen.

Phase 2

- During this round, a position would be reversed, with an ambiguous site setting their transmitting power to 1 before broadcasting to nearby reference nodes and waiting for their responses.
- Power transmission was set to 1 by the ambiguous sites (12 dBm). Up to 130 meters is the broadcast and receiving range thanks to `[xbee802.setPowerLevel(1)]`.
- Unidentified nodes begin broadcasting communications that include their IDs and Mac addresses. utilizing respective (X, Y), as well as Identity data, all reference sites that have received a broadcast signal will respond to the unknown sites.
- Ambiguous sites will get these signals as well as only utilize these new entry sites at one radius somewhere between 100 to 130 meters [(accepted references) - (Phase 1 connections)].

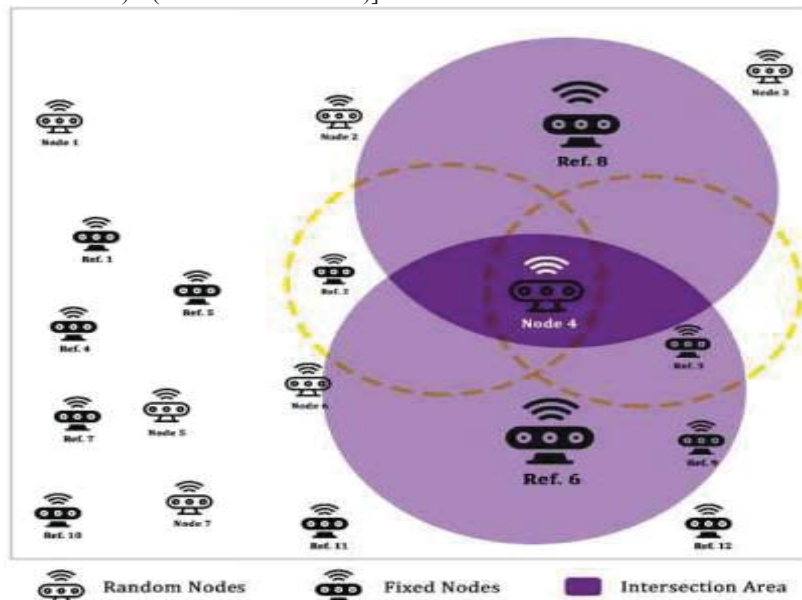


Fig. 8. The Suggested Destinations in the Phase 1 and Phase 2 interconnection of the Chosen References Node Range.

- This quantity of potential Destinations determined by Phase 1 will be reduced using the traditional distance-between-two-points algorithm, and the distance between each potential site determined by Phase 1 as well as all sources determined by Phase 2 shall be computed.

- It will be stored inside a potential destination of Phase 2 $[[[]] = [x_2 \ y_2 \dots \ x_{n2} \ y_{n2}]$ if $(R_1 = 100)$ Square root $[(x \text{ round1} - x \text{ reference phase 2})^2 + (y \text{ phase 1} - y \text{ reference phase 2})^2] = (R_2 = 130)$. If not, it will be deleted. (Observe Figure 10).
- This round's locations will be used to determine the mean of the possible sites.

Phase 3

- Up to 180 meters of transmission and reception are possible with the energy transfer set by ambiguous site to $2 = 14$ dBm $[xbee802.setPowerLevel(2)]$.
- Unidentified nodes begin broadcasting messages with their ID and Mac address.
- Referencing sites will respond to unidentified sites with their (X, Y), and Identity information if they receive the broadcast signals.
- At a distance of between 130 and 180 meters, Ambiguous sites are only going to utilize new entry sites and get the information. As shown in the image, $[(\text{received reference}) - (\text{Round 2 references}) - (\text{Phase 1 references})]$.
- Each possible location from Phase 2 and each reference from Phase 3 will be measured in relation to one another. This will reduce the amount of Phase 2 origins that are feasible.
- The location will be saved in a probable location of Phase 3 $[[[]] = [x_3 \ y_3 \dots \ x_{n3} \ y_{n3}]$ if $(R_2 = 130)$ Sqrt $[(x \text{ phase2} - x \text{ reference phase 3})^2 + (y \text{ phase 2} - y \text{ reference phase 3})^2] = (R_3 = 180)$. If not, it will be deleted. (Refer to Figure 11).
- This round's destinations will be used to determine the mean of the possible sites.

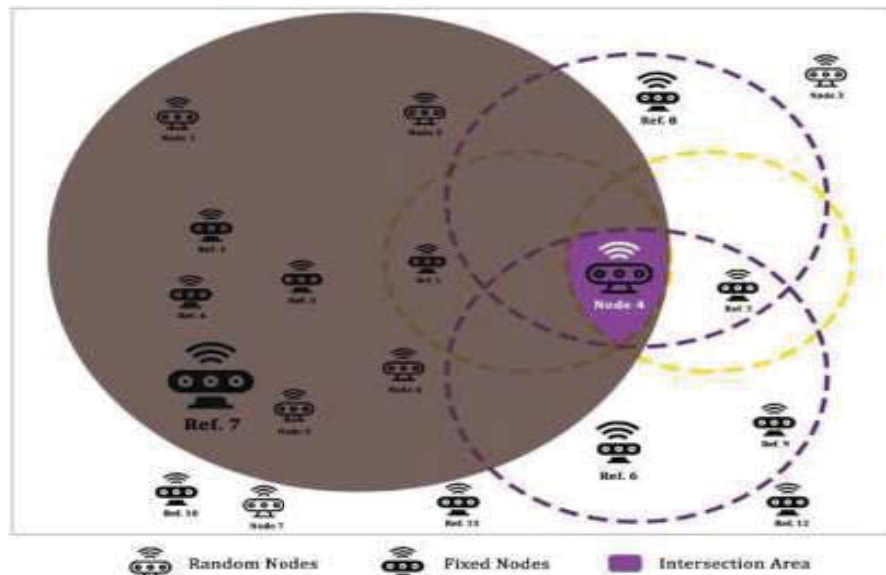


Fig. 9. The Suggested Destinations in the Phase 1, Phase 2 and Phase 3 interconnection of the Chosen References Node Range.

Phase 4

- Ambiguous sites set the energy transmission to $3 = 16$ dBm $[xbee802.setPowerLevel(3)]$, which allows for up to 260 meters of transmission and reception.
- Ambiguous sites begin transmitting broadcast messages containing their Mac address and Identity.
- As soon as you get broadcast signals, any Reference node will respond to any unknown nodes by providing their (X, Y), as well as Identity data.
- At a distance of between 180 and 260 meters, Ambiguous sites would get these signals and solely employ the fresh reference sites. Phase 3 references, Phase 2 references, and (received reference) (Phase 1 references)
- Each prospective Phase 3 site's distance from every reference coming from phase 4 will be calculated. The distance formula would be applied to cut down on the number of Phase 3's potential places.
- If $(R_3 = 180)$ Sqrt $[(x \text{ phase 3} - x \text{ reference phase 4})^2 + (y \text{ phase 3} - y \text{ reference round 4})^2] = (R_4 = 260)$, then a prospective destination will save the address of Phase 4 $[[[]] = [x_4 \ y_4 \dots \ x_{n4} \ y_{n4}]$. It will be erased if not.

- The node is now prepared to shift its role and assist other unallocated nodes in localizing themselves by being suggested as the average of potential nodes' estimated locations. The method is summarized in the flowchart below. (See Figure 12).

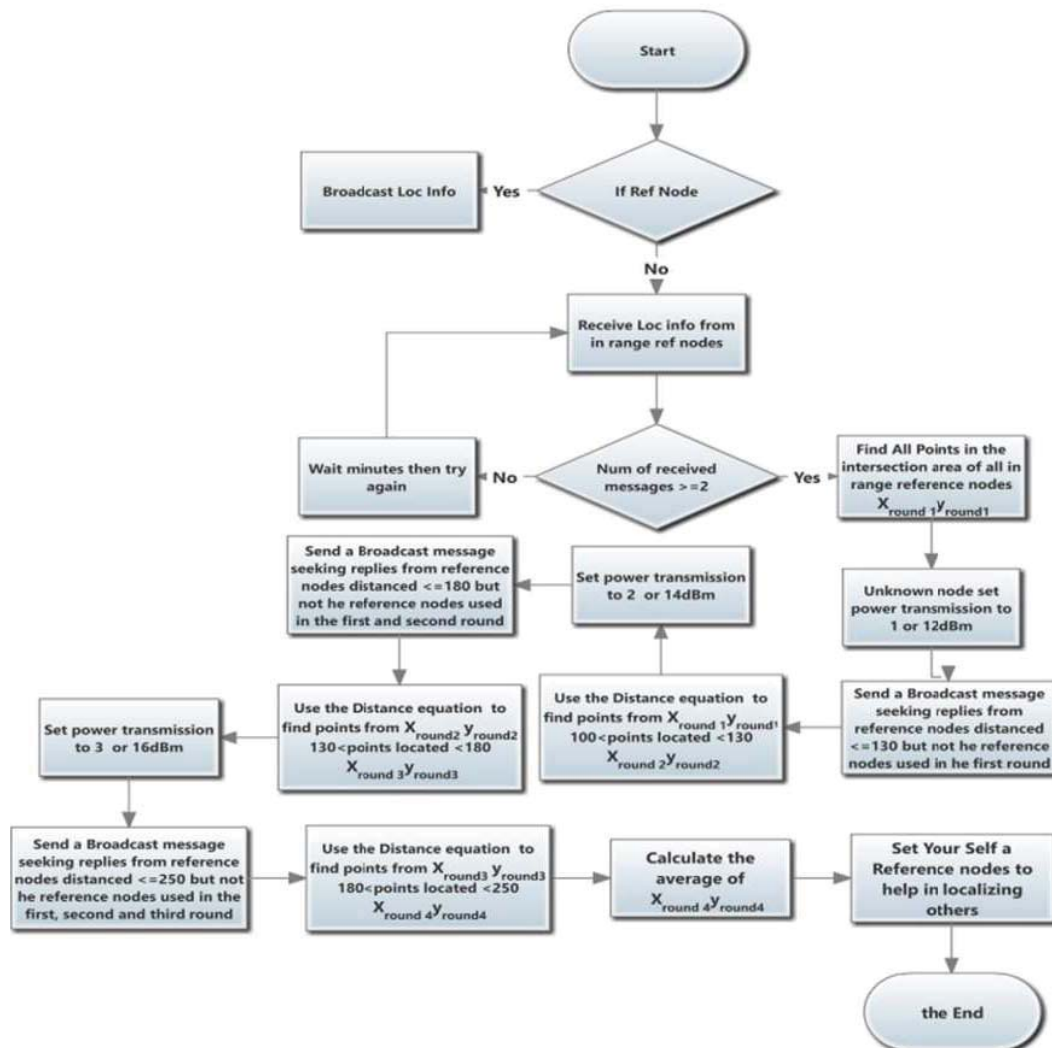


Fig. 10. Flow Chart of the technique.

1. **Start**
2. **Ref Nodes:** Set power to first level.
3. **Ref Node:** Send broadcast message.
4. **Receiving Xnode:** Calculate intersection points
5. **While** power \leq max level
 - **Receiving Xnode:** Set power to the next level.
 - **Receiving Xnode:** Send a broadcast message.
 - **Receiving Ref Nodes:** Set power to the next level.
 - **Receiving Ref Nodes:** Reply with a unicast.
 - **Receiving Xnode:** Find the new intersection points.
6. **End While**
7. **End.**

Fig. 11. Pseudo Code.

5. Results and Discussion

We've conducted tests in a woodland setting. The communication range was manually measured within a 400×400 m region where the nodes were dispersed in order to determine whether XBee power transfer can be configured to 0, 1, 2, or 3. The positioning of the 20 sites. (Observe Fig. 14).

Four sets of experiments were conducted to assess how the suggested strategy affected localization accuracy. Each round's outcomes are described (see Table 3).

Table 3 Real nodes and their measured locations after four iterations.

Node Type	Node ID	Actual X	Actual Y	Round 1			Round 2			Round 3			Round 4		
				X	Y	# of Possible Locations	X	Y	# of Possible Locations	X	Y	# of Possible Locations	X	Y	# of Possible Locations
REF	1	100	100	-	-	-	-	-	-	-	-	-	-	-	-
REF	2	200	100	-	-	-	-	-	-	-	-	-	-	-	-
REF	3	300	100	-	-	-	-	-	-	-	-	-	-	-	-
REF	4	100	200	-	-	-	-	-	-	-	-	-	-	-	-
REF	5	200	200	-	-	-	-	-	-	-	-	-	-	-	-
REF	6	300	200	-	-	-	-	-	-	-	-	-	-	-	-
REF	7	100	300	-	-	-	-	-	-	-	-	-	-	-	-
REF	8	200	300	-	-	-	-	-	-	-	-	-	-	-	-
REF	9	300	300	-	-	-	-	-	-	-	-	-	-	-	-
Random	10	130	140	150	150	37	*	*	*	153	153	20	130	140	1
Random	11	190	150	200	150	127	200	150	9	195	155	2	190	150	1
Random	12	260	140	250	150	37	*	*	*	246	153	18	246	153	18
Random	13	120	270	142	257	48	120	280	10	113	266	3	113	286	3
Random	14	270	270	250	250	37	*	*	*	247	247	23	270	270	1
Random	15	334	147	300	150	127	*	*	*	346	152	21	340	150	15
Random	16	315	239	300	250	127	298	252	30	316	233	3	*	*	*
Random	17	145	62	150	100	127	*	*	*	152	53	21	150	60	15
Random	18	194	74	150	100	127	182	91	7	186	88	5	186	88	5
Random	19	229	347	250	300	127	*	*	*	247	346	21	250	340	15
Random	20	267	238	252	257	48	280	280	10	286	266	3	286	266	3

According to the results (see Table 4), the node results of potential sites were decreased in each cycle, which also decreased the likelihood that the expected location would be incorrect.

We observed that errors in rounds 1 and 2 ranged from 10 to 51 metres, round 3 from 7 to 26 metres, and round 4 from 0 to 22 metres. It's important to note that not all nodes locate themselves precisely and without making any mistakes. (View Figures 15 and 16). See table 5 to compare our findings to those of other approaches.

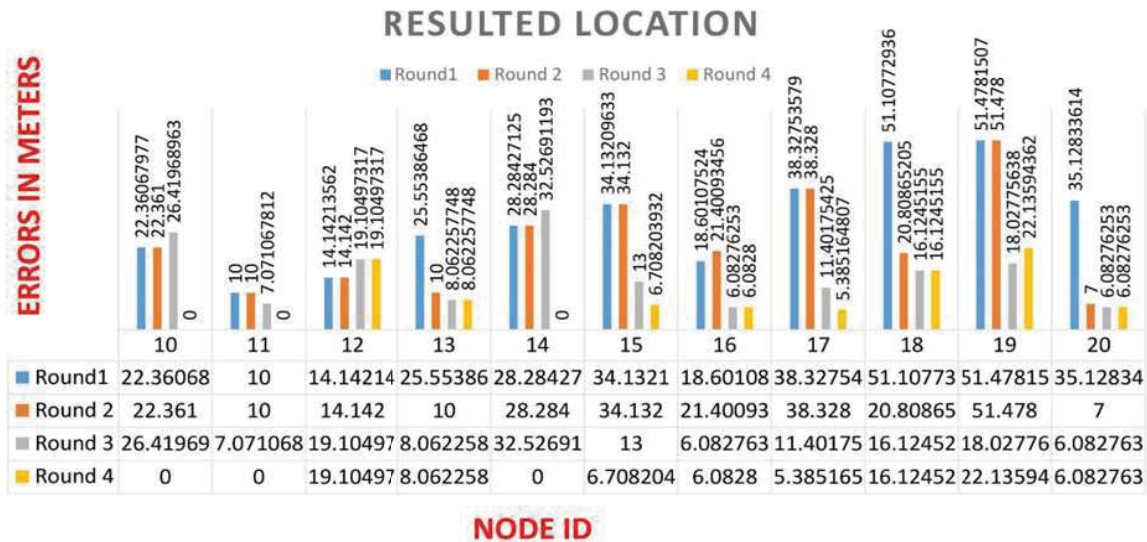


Fig. 12. The typical meters-per-round inaccuracy.

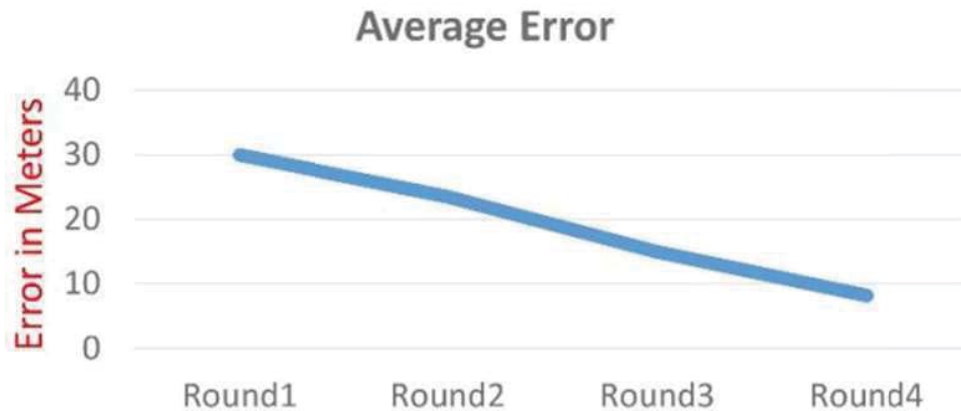


Fig. 13. The increase in the typical meter inaccuracy for each round.

6. Conclusions

Accurate localisation is one of the most crucial elements in the majority of WSN applications. Trilateration, Triangulation, AOA, TDOA, TOF, GPS, TOA and TOF are common localization methods. The following issues with these methods could arise: 1. Both the cost and node life duration of GPS are excessively high. 2. RSSI is used by the majority of ranging techniques to determine distances. In actual situations, the signal strength does not always decrease as the range between transmitter and receiver increases, though. These RSSI readings can change irregularly over time, even when there is a consistent distance between the devices. 3. The TOA and TOF approaches cannot be used because WSN nodes powered by the only exact timing that Waspnote and Arduino CPUs can provide is millisecond resolution. Only one method that works in short-range communication networks without additional node localization components is the communication ranging technique. The range technique was employed several times in the suggested method to find nodes. By leveraging products that were already on the market, this method improved accuracy to the point where it occasionally neared 100%. Additionally, with each round, the number of potential locations shrank. The suggested method can be applied to external uses. Let's select a woodland region as the location for the trail, which has a variety of challenges. The method has the benefits of price cutting as well as increasing network lifespan even in environments with challenging radio propagation and in inhospitable locations. Moreover, it functions with lots of nodes.

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